

# Linguistic tone is related to the population frequency of the adaptive haplogroups of two brain size genes, *ASPM* and *Microcephalin*

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Edited by Henry C. Harpending, University of Utah, Salt Lake City, UT, and approved April 12, 2007 (received for review December 7, 2006)

The correlations between interpopulation genetic and linguistic diversities are mostly noncausal (spurious), being due to historical processes and geographical factors that shape them in similar ways. Studies of such correlations usually consider allele frequencies and linguistic groupings (dialects, languages, linguistic families or phyla), sometimes controlling for geographic, topographic, or ecological factors. Here, we consider the relation between allele frequencies and linguistic typological features. Specifically, we focus on the derived haplogroups of the brain growth and development-related genes *ASPM* and *Microcephalin*, which show signs of natural selection and a marked geographic structure, and on linguistic tone, the use of voice pitch to convey lexical or grammatical distinctions. We hypothesize that there is a relationship between the population frequency of these two alleles and the presence of linguistic tone and test this hypothesis relative to a large database (983 alleles and 26 linguistic features in 49 populations), showing that it is not due to the usual explanatory factors represented by geography and history. The relationship between genetic and linguistic diversity in this case may be causal: certain alleles can bias language acquisition or processing and thereby influence the trajectory of language change through iterated cultural transmission.

learning biases | tone language | linguistic typology | cultural transmission

Human populations are diverse both genetically and linguistically, through interpopulation differences in allele frequencies (1–3) and in the variety of languages and dialects they speak (4). In general, any relationship between these two types of diversity merely reflects geography and past demographic processes, not genetic influence on language behavior (1, 2, 5–8). It is indisputable that normal infants of any genetic makeup can learn the language(s) they are exposed to in the first years of life, so we can assume with considerable confidence that there are no “genes for Chinese.”

Nevertheless, it is well accepted that there is widespread inter-individual variation in many aspects relevant for language [developmental delays, differences in second-language learning aptitude, discrimination between foreign speech sounds (9), recognition of words in noise (10), and differences in short-term phonological memory correlated with different syntactic processing strategies (11)]. It is also accepted that this variation can be partially attributed to genetic factors, most probably through a “many genes with small effects” model including both generalist and specialist genes (12–15). There are also heritable aspects of brain structure in general, and language-related areas in particular (16–21).

It is therefore likely that there are heritable differences of brain structure and function that affect language acquisition and usage. These differences may have no obvious behavioral consequences in the nonclinical population; under ordinary circumstances, all normal speakers and hearers perform “at ceiling” on many language-related tasks (10). Moreover, no one doubts that all normal children acquire the language of the community in which they are reared. Nevertheless, if differences in language and speech-related capac-

ities are variable and heritable and if the genes involved have interpopulation structure, it is likely that populations may differ subtly in some of these aspects, and that differences between populations could influence the way languages change through cultural evolution over time.

It is generally acknowledged (22) that the process of language acquisition plays a major role in historical language change: language acquirers construct a grammar based on the language they hear around them, but the constructed grammar is not necessarily identical to that of their models, and the cumulative effect of such small differences over generations leads to language change. It follows that cognitive biases in a population of acquirers could influence the direction of language change across generations. These biasing effects could result in linguistic differences between populations, producing nonspurious (causal) correlations between genetic and linguistic diversities. Computer simulations (23, 24) support the idea that such biases could influence the structure of languages emerging over many generations of cultural change, and mathematical models (25) suggest that, under appropriate conditions, extremely small biases at the individual level can be amplified by this process of cultural transmission and become manifest at the population level.

## Linguistic Tone

We propose that the linguistic typology of tone is affected by such a bias. Human languages differ typologically in the way they use voice fundamental frequency (pitch). All languages use consonants and vowels to distinguish one word or grammatical category from another, but, in addition, so-called “tone languages” (e.g., Chinese) use pitch for this purpose as well, whereas “non-tone languages” (e.g., English) use pitch only at sentence level (to convey emphasis, emotion, etc.) (26). In tone languages, that is, pitch is organized into tone phonemes that are functionally comparable with consonant and vowel phonemes. Tone languages are the norm in sub-Saharan Africa and are very common in continental and insular southeast Asia. They are rare in the rest of Eurasia, North Africa, and Australia. They are relatively common in Central America, the Caribbean, and the Amazon basin, and occur sporadically elsewhere among the aboriginal languages of the Americas (27).

Author contributions: D.D. and D.R.L. designed research; D.D. and D.R.L. performed research; D.D. and D.R.L. analyzed data; D.D. was primarily responsible for genetic data and statistical analysis; D.R.L. was primarily responsible for language data; and D.D. and D.R.L. wrote the paper.

The authors declare no conflict of interest.

This article is a PNAS Direct Submission.

Abbreviations: *ASPM*-D, the derived (adaptive) haplogroup of the gene *ASPM*; *MCPH*-D, the derived (adaptive) haplogroup of the gene *Microcephalin*; DV, dependent variable; IV, independent variable.

See Commentary on page 10755.

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The vast majority of the world's languages are unambiguously either tonal or not (27), but a few languages (e.g., Japanese, Swedish/Norwegian, and Basque) are typologically intermediate, and it is well established that languages can lose or acquire tone through ordinary historical change (28). More strikingly, there are cases showing that the difference between "tonal" and "nontonal" languages can actually be quite subtle, such as the existence of closely related (even mutually intelligible) languages and dialects of which some are tonal and some are not. The best described such cases are Kammu in Laos (29) and various Alaskan Athabaskan languages (30). In both cases, the phonological interpretation of pitch differences associated with obstruent voicing (Kammu) or coda glottalization (Athabaskan) is ambiguous in a way that could drive language change: specifically, these differences might be perceived by an acquirer either as part of a system of contrastive tones, or as allophonically conditioned accompaniments of glottalized or voiced obstruent phonemes. If, as we propose, tone is affected by some form of acquisition or processing bias, we might expect that it would manifest itself in cases like these. Although the exact nature of the bias is currently unclear, it is plausible that it might involve a propensity to favor linguistic structures in which elements such as phonemes and morphemes are strictly linearly ordered rather than (as is the case with tone) simultaneous or formally unordered.

A recent series of studies conducted by Wong and colleagues (31, 32) seems to point to interindividual differences in tone learning and associated neural correlates. Adult speakers of a nontonal language (English) were presented with an artificial language learning task involving lexical tonal distinctions, and it was found that they tend to form two groups, referred to as "successful" and "less successful" learners. A later study by the same team (P. C. M. Wong, personal communication), focusing on the relationship between the anatomy of the primary auditory cortex and linguistic tone learning, found that the successful learners showed greater volume of left, but not right, Heschl's Gyrus, especially for gray matter. Although this correlation could be entirely due to environmental effects of previous experience, it could also point to a genetic component. Interestingly, there are suggestions in the literature concerning the heritability of musical pitch processing (33) and the genetics of absolute pitch (34), and, whereas the relationship between linguistic and musical/absolute pitch is by no means simple (35), these studies are certainly consistent with the proposal of a genetic bias affecting linguistic tone.

### **ASPM and Microcephalin**

*ASPM* (*MCPH5*, 1q31) and *Microcephalin* (*MCPH1*, 8p23) are two genes involved in brain growth and development (36–38). Deleterious mutations of both *ASPM* and *Microcephalin* are involved in recessive primary microcephaly (38–40), together with at least four other loci identified to date (39, 41). During embryogenesis, the neuroepithelial cells, found around the telencephalic ventricle (42), undergo two types of division: symmetric, producing two neuroepithelial cells, or asymmetric, producing a neuroepithelial cell and a neuronal precursor (43), which migrates toward its final position in the cortex (42). The type of cell division depends on the orientation of the mitotic spindle relative to the apical–basal axis (43). It has been suggested (44) that a change in the number of symmetric divisions will dramatically alter brain size, given that each such division potentially doubles the final number of neurons. Both *ASPM* and *Microcephalin* are involved in cell–cycle regulation (45–48), and their deleterious mutations impact on the number of such symmetric divisions. It has been suggested that *ASPM* insures the maintenance of the perpendicular position of the mitotic spindle in the neuroepithelial cells, a very difficult task given their extremely elongated shape (43), which cannot be correctly accomplished by the truncated proteins associated with

the deleterious mutations. Moreover, a recent report (49) suggests a putative ciliary function for *ASPM*, pointing to an influence on neuronal migration, mediated by cerebrospinal fluid flow. For *Microcephalin*, the mechanism seems to be represented by the failure of the truncated protein to protect the neuroepithelial cells against DNA repair defects, leading to excessive apoptosis (39).

For both genes, "derived" haplogroups have been identified (the G allele for the A44871G polymorphism for *ASPM*, and the C allele for the G37995C polymorphism for *Microcephalin*) (36, 37). These haplogroups will be denoted as *ASPM-D* and *MCPH-D*, respectively. Their ages are estimated at 5.8 thousand years (ky) [95% confidence interval (C.I.); 0.5–14.1 ky] and 37 ky (95% C.I.; 14–60 ky), respectively, both showing signs of positive selection and a marked geographic structure (36, 37). *ASPM-D* reaches high frequencies in Central and Western Asia, Europe and North Africa, as well as in Papua-New Guinea (but there are reasons to suspect contamination; see *Discussion*) and very low frequencies in East Asia, sub-Saharan Africa, and the Americas (see map in ref. 36). *MCPH-D* is very frequent in Asia, Europe, and the Americas, moderately frequent in North and East Africa, South-East Asia, and Oceania (see comment on Papua-New Guinea), and very rare in Central, Western, and South sub-Saharan Africa (see map in ref. 37). Moreover, both genes show signs of accelerated evolution in the human lineage (approximately two favorable mutations per million years; ref. 38). The claim that the distribution of *ASPM-D* and *MCPH-D* is the result of positive selection has recently been challenged (50) but arguably remains the best explanation (51).

The phenotypic effects of the derived haplogroups of *ASPM* and *Microcephalin* are not yet known but arguably do not include gross phenotypic alterations: the derived haplogroups are apparently not involved in variations in intelligence (52), brain size (53), head circumference, general mental ability, social intelligence (54), or the incidence of schizophrenia (55). We propose that their effects involve subtle differences in the organization of the cerebral cortex, with cognitive consequences including linguistic biases in the processing and acquisition of linguistic tone. More specifically, based on the suggestions in ref. 43, it is highly possible that *ASPM-D* alters the orientation of the mitotic spindle dependent on local conditions in the precursors of language areas, leading to the emergence of the suggested bias. Moreover, it is plausible that *MCPH-D* contributes to this process by influencing the number of symmetric divisions. One could envisage a hypothetical scenario whereby the changes induced by *MCPH-D* are enhanced by *ASPM-D* through a modification of the precise maintenance of the orientation of the mitotic spindle during the development of specific language-related areas.

### **Hypothesis**

These considerations led us to hypothesize a relationship between the distribution of tone languages and the geographical structure of *ASPM-D* and *MCPH-D*. Those areas of the world where the new alleles are relatively rare also tend to be the areas where tone languages are common. As previously discussed, the effects of *ASPM-D* and *MCPH-D* on brain structure and functioning remain largely hypothetical, but it is entirely plausible that they influence the cognitive capacities involved in processing phonological structures and thereby lead to linguistic biases of the type suggested above.

In the present study, we performed statistical tests of this hypothesis on the basis of a large database comprising 983 alleles and 26 linguistic features collected for 49 world populations (see *Materials and Methods*), controlling for geographical and historical factors. We considered linguistic features rather than linguistic groupings (dialects, languages, linguistic families, or phyla), because our hypothesis concerns specifically the interaction between linguistic typological diversity and population genetic diversity. We found that, in general, the relationship between these two diversities is fully explained by geographical and historical factors, whereas, in

the specific case of tone, *ASPM-D*, and *MCPH-D*, there is an important and significant correlation between their distributions even after controlling for geography and history. Therefore, we propose that this relationship is causal; that is, the genetic structure of a population can exert an influence on the language(s) spoken by that population. Further experimental support is required, but these findings suggest a fundamental direction for future research targeted at understanding the complex relationship between genetic factors, cultural evolution, and linguistic phenomena.

## Results

In the following, we have systematically applied Holm's multiple comparisons correction (56), and the reported *P* values are adjusted. All of the statistical analyses used R (57).

**The Relationship Between Linguistic Features and Alleles.** The first aspect of the hypothesis concerns the existence of a relationship between the linguistic feature of tone and the derived haplogroups of *ASPM* and *Microcephalin*. We tested this aspect by comparing the strength of the relationships between tone and *ASPM-D* and between tone and *MCPH-D* with the distribution of the relationships between all 26 linguistic features and all 983 genetic markers in our database. Specifically, we computed the distribution of the resulting values of Pearson's correlation coefficients, *r*, and found it to be normal for all pairs of linguistic features ( $n = 325$ , mean = 0.012, SD = 0.274), all pairs of alleles ( $n = 482,653$ , mean = 0.024, SD = 0.225), and all pairs of linguistic features and alleles ( $n = 25,558$ , mean = -0.006, SD = 0.218). This result shows that, in general, linguistic features do not correlate with alleles. Focusing on the distribution of Pearson's *r* for all pairs of linguistic features and alleles, we found that the correlations between tone and *ASPM-D* and between tone and *MCPH-D* are both highly significant (tone and *ASPM-D*:  $r = -0.53$ ,  $P = 9.63 \times 10^{-5}$ ; tone and *MCPH-D*:  $r = -0.54$ ,  $P = 7.22 \times 10^{-5}$ ) and their values are in the top 1.5% of the empirical distribution of correlations.

This result shows that, taken individually, tone and *ASPM-D* and tone and *MCPH-D* are highly significantly correlated and the strength of their relationship is >98.5% of all of the 25,558 correlations between linguistic features and alleles in our database.

**The Relationship Between Linguistic Features and Pairs of Alleles.** The second aspect of our hypothesis concerns the relationship between tone and both *ASPM-D* and *MCPH-D*, which we tested using a logistic regression approach (58). We computed the logistic regressions of all linguistic features [as the dependent variables (DVs)] on all pairs of alleles [as the independent variables (IVs)] ( $n = 11,582,690^*$ ), and their distribution is heavily skewed toward poor fit, as expected. However, the logistic regression of the DV tone on the IVs *ASPM-D* and *MCPH-D* is both very good (Nagelkerke's  $R^2 = 0.528$ , 73% correct classification; Intercept: estimate = 4.478, SE = 1.843,  $P = 0.015$ ; *ASPM-D*: estimate = -7.170, SE = 2.767,  $P = 0.010$ ; *MCPH-D*: estimate = -4.952, SE = 2.217,  $P = 0.026$ ) and in the top 2.7% of the empirical distribution of the logistic regressions. We also tested the effects of the interaction between *ASPM-D* and *MCPH-D* on tone (58), by performing the logistic regression of the DV tone on the IVs *ASPM-D*, *MCPH-D*, and *ASPM-D*\**MCPH-D*, but the interaction term is nonsignificant ( $P = 0.224$ ) and the new model does not perform better [ $\chi^2(1) = 1.848$ ,  $P = 0.174$ ].

This result shows that tone and the pair *ASPM-D*/*MCPH-D* are highly significantly related and the strength of their relationship is >97.3% of all of the 11,582,690 converged logistic regressions between linguistic features and pairs of alleles in our database.

**Controlling for Geographical and Historical Factors.** To control for the effects of geography and shared linguistic history on our results, we compared geographic, genetic, typological linguistic, and historical linguistic distances between all pairs of populations in the sample. The land (geographic) distances are represented by great circle distances for pairs of populations on the same continent, with intercontinental paths forced through specific connection points (Damascus for Africa/Eurasia and Bangkok for Melanesia/Eurasia). The genetic distances are represented by Nei's *D* (59). For any set of linguistic features, the typological linguistic distance represents a generalized Euclidean distance over the space of these linguistic features (see *Materials and Methods*). The historical linguistic distance is based on the linguistic relatedness given by historical linguistic classifications, as follows (60): the value is 1 if the populations speak the same language, 2 if they speak languages belonging to the same branch of a linguistic family, 3 if they speak languages from different branches of the same linguistic family, and 4 if they speak languages not demonstrably related. Historical linguistic judgments are based on the classification in ref. 4 and exclude controversial items.

We studied the relationships between these distances using Mantel (partial) correlations (61):  $r = 0.509$ ,  $P < 0.001$  (geographic vs. genetic);  $r = 0.283$ ,  $P < 0.001$  (geographic vs. typological linguistic);  $r = 0.162$ ,  $P = 0.011$  (genetic vs. typological linguistic) and  $r = 0.021$ ,  $P = 0.407$  (genetic vs. typological linguistic, while controlling for geographic distances). In general, therefore, the (weak) correlations between genetic and typological linguistic diversities can be accounted for by geography, confirming that, generally, there is no direct influence of genes on language behavior (2, 5). Because we are referring to typological linguistic diversity rather than the historically based linguistic diversity of Cavalli-Sforza and coworkers (1), our results of a general lack of correlation between linguistic and genetic diversities do not contradict their findings.

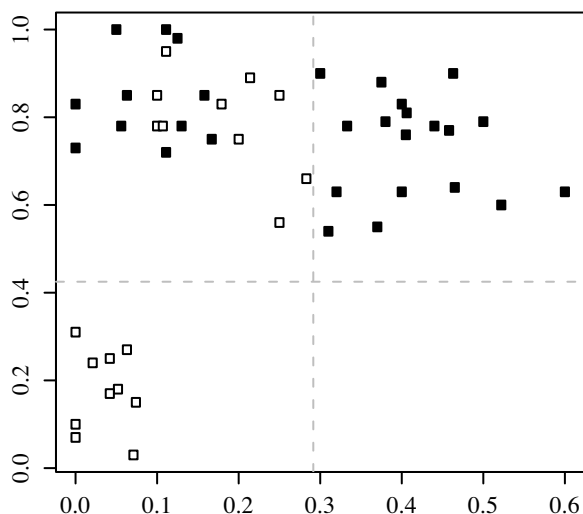
Individually, the Mantel correlation with geography for tone is  $r = 0.169$ ,  $P = 0.015$ ; for *ASPM-D*,  $r = 0.074$ ,  $P = 1.000$  (because of Holm's multiple comparisons correction; ref. 56); and for *MCPH-D*,  $r = 0.543$ ,  $P < 0.001$ . Each of tone, *ASPM-D*, and *MCPH-D* have low but significant spatial autocorrelations (62): Moran's *I* (63) is 0.178, 0.164, and 0.121, and Geary's *c* (64) is 0.634, 0.438 and 0.718, respectively,  $P < 0.001$  for all, suggesting that, potentially, geographical factors might explain the observed relationship. However, the (partial) Mantel correlation between tone and the pair *ASPM-D*/*MCPH-D* is  $r = 0.333$ ,  $P < 0.001$ , and, when controlling for geography, it decreases only slightly and still remains highly significant,  $r = 0.291$ ,  $P = 0.003$ , showing that geography is not a good explanation for our empirical findings.

Tone, *ASPM-D*, and *MCPH-D* tend to be much more similar inside than across linguistic families [the linguistic and genetic distances between populations speaking languages of the same families are smaller than across families: random permutations test (65),  $P < 0.001$ ], suggesting that the shared linguistic history might explain the observed relationship between them. However, when controlling for the historical linguistic distances, the partial Mantel correlation between tone and the pair *ASPM-D*/*MCPH-D* remains important and highly significant ( $r = 0.271$ ,  $P < 0.001$ ), showing that the relationship cannot be fully explained in this manner.

Moreover, when controlling simultaneously for geography and shared linguistic history, the second-order partial Mantel correlation between tone and the pair *ASPM-D*/*MCPH-D* actually increases slightly and is highly significant ( $r = 0.283$ ,  $P < 0.001$ ), suggesting not only that geographical factors and shared linguistic history do not explain the hypothesized relationship, but that the linguistic history represents a suppressor variable (58) on this relationship.

Fig. 1 represents the distribution of linguistic tone as a function of the population frequency of *ASPM-D* and *MCPH-D*. Open squares stand for the tonal languages, and

\*This is the number of logistic regressions for which the algorithm converged.



**Fig. 1.** Linguistic tone versus the population frequency of the adaptive haplogroups of *ASPM* and *Microcephalin*. The horizontal axis represents the frequency of *ASPM-D*, whereas the vertical axis represents the frequency of *MCPH-D*. Filled squares represent nontonal languages and open squares tonal languages. Gray dashed lines correspond to 0.292 *ASPM-D* and 0.425 *MCPH-D*. See last paragraph of *Results* for details.

their distribution corresponds to low frequencies of *ASPM-D* (lower than  $\approx 0.29$ ), whereas filled squares stand for nontonal languages, and their distribution corresponds to high frequencies of *MCPH-D* (higher than  $\approx 0.42$ ). Strikingly, in the bottom-left quadrant, there are only tonal languages and, in the top-right quadrant, only nontonal languages whereas, in the top-left quadrant, there is an even distribution of tonal and nontonal languages (10:11). There are no populations in our sample occupying the bottom-right quadrant. This figure illustrates the (probabilistic) predictions of our model concerning the tonality of a language given the frequency of *ASPM-D* and *MCPH-D* in that population. These predictions are corroborated by the five American populations not included in the analysis, which have low frequencies of *ASPM-D* and high frequencies of *MCPH-D*; as expected, their languages are both tonal and nontonal. (We exclude the Papuan population from further consideration, because it seems likely to be unreliable because of contamination; see *Materials and Methods*). A very important test case for our model would be provided by Australia, because the Australian languages are nontonal; however, obtaining reliable genetic samples seems very difficult.

## Discussion

In this article, we formulated and tested the hypothesis of a nonspurious correlation between linguistic tone and the derived haplogroups of two genes involved in brain growth and development, *ASPM* and *Microcephalin*. In so doing, we have also introduced a previously undescribed methodology for studying the relationship between genetic and linguistic diversities. Although we are well aware that a correlational approach cannot by itself prove causality, we have shown that our hypothesis is supported by the currently available data. Specifically, we have found that the negative correlation between tone and the population frequency of *ASPM-D* and *MCPH-D* cannot be explained by historical and geographical factors, thus strengthening the claim of a causal relationship between them. As noted in the introduction, we propose that the causal relation is mediated by a cognitive bias relevant to the processing and acquisition of tone.

We may summarize the structure of the proposed genetic influence on the distribution of linguistic tone in three necessary components or causal steps: from interindividual genetic differences to differences in brain structure and function, from these differences in brain structure and function to interindividual differences in language-related capacities, and, finally, to typological differences between languages. The first component is represented by the proposed effects of *ASPM-D* and *MCPH-D* on brain structure and function, including the brain areas involved in linguistic tone. The second component involves interindividual differences in the acquisition and/or the processing of tone, which are supported by several recent findings. The last component, probably the best supported to date, relies on the process of cultural transmission of language across generations, which can, in the right circumstances, amplify small individual biases to influence the trajectory of language change. We assume that any such bias is very small at the individual level and becomes manifest only at the population level through the process of cultural transmission. We also assume that the bias is probabilistic in nature and that many other factors, including language contact and history, also govern the process of language change and affect its outcome. Our findings therefore do not support any racial or deterministic interpretation. Finally, note that this bias could be either for or against tone, but the fact that nontonicity is associated with the derived haplogroups (Fig. 1) suggests that tone is phylogenetically older and that the bias favors nontonicity. The bias is presumably a selectively neutral byproduct of the two derived haplogroups, not connected to the selective pressures on them, because there is no evidence that tone itself confers any advantage or disadvantage on speakers. We cannot, of course, rule out the scenario whereby the natural selection detected for these haplogroups is partially due to their linguistic effects.

The correlation reported here represents a plausible and previously undescribed case in which differences in population genetic structure partially account for linguistic differences. This finding warrants future experimental work, which will help test and refine the hypothesis of a causal effect. The artificial language learning paradigm of Wong and colleagues (31) offers a solid framework for testing whether the existence of individual biases in the acquisition and processing of linguistic tone is influenced by the presence or absence of *ASPM-D* and *MCPH-D*. A study of the effects of these derived haplogroups on other language-related capacities, including phonological working memory or pitch tracking, is also warranted. Additionally, research is clearly needed on the phenotypic effects of these haplogroups on brain structure. Depending on the outcome of such experimental work, the results reported here could lead to a profound change in our understanding of the interactions between genetic diversity and our higher cognitive capacities, by bridging the gap between interindividual and interpopulation diversities. They also represent a solid foundation for gradual, accretionary models of language evolution and suggest a hitherto unsuspected mechanism driving language change.

## Materials and Methods

**Populations.** The 49 populations used in this study were selected from the 59 populations in refs. 36 and 37 based only on genetic and linguistic data availability. The Americas were too poorly sampled for their genetic and linguistic diversity, so that the five American populations have been excluded from the analysis but have been used as a test case. “Orogen” is probably a misspelling of “Oroqen.” The populations have been identified geographically, linguistically, and genetically by using information from various sources (refs. 4 and 66–68; Maps Of World: [www.mapsofworld.com/lat\\_long/index.html](http://www.mapsofworld.com/lat_long/index.html), accessed April 17, 2007). Because of systematically missing genetic information (see below), four African populations were eliminated (Masai, Sandawe, Burunge, and Zime). Also, Papuan was eliminated because of its ambiguity and the high

probability of contamination, suggested by its low genetic similarity to neighbors, but high to Europe. The NAN Melanesian (Non-Austronesian Melanesian) population is very poorly specified in refs. 36 and 37, but it most probably represents (66, 67) the Naasioi of Bougainville, Papua New Guinea. The 49 populations are as follows<sup>8</sup>: Southeastern and Southwestern Bantu, San (naq), Mbuti Pygmy (efe), Turu (rim), Northeastern Bantu (kik), Biaka Pygmy (axk), Bakola Pygmy (gyi), Bamoun (bax), Yoruba (yor), Mandenka (mnk), Mozabite (mzb), Druze (apc), Palestinian (ajp), Bedouin (ayl), Hazara (haz), Balochi (bgp), Pathan (pst), Burusho (bsk), Makrani (bcc), Brahui (brh), Kalash (kls), Sindhi (snd), Hezhen (gld), Mongola (mvf), Daur (dta), Orogen (orh), Miaozi (hmy), Yizu (yif), Tujia (tji), Han (cmn), Xibo (sjo), Uygur (uig), Dai (tdd), Lahu (lhu), She (shx), Naxi (nbf), Tu (mjg), Cambodian (khn), Japanese (jpn), Yakut (sah), NAN Melanesian (nas), French Basque (eus), French (fra), Sardinian (src), North Italian (vec), Tuscan (ita), Orcadian (sco), Russian (rus), and Adygei (ady).

**Genetic Data.** For each of the 49 populations, frequency and positional information was gathered about *ASPM-D* and *MCPH-D* (36, 37), as well as  $\approx 133$  alleles from the ALFRED database (66, 67) and 1,029 from the HDPG data set (69), the only criterion being that frequency information is available for at least 44 of the 49 populations (the vast majority, except *ASPM-D* and *MCPH-D* are short tandem repeats). Positional information was obtained from the UniSTS Project (70), and for 50 alleles no such information could be retrieved. Moreover, 124 pairs were duplicated between the two databases, and 9 were deleted because they introduced systematic missing data in sub-Saharan Africa. After these deletions, 981 alleles were retained.

Because genetic information is missing for most sub-Saharan populations, for the five populations speaking languages belonging to the Narrow Bantu branch of the Niger-Congo linguistic family (4) (Southeastern and Southwestern Bantu, Turu, Northeastern Bantu, Bakola Pygmy, and Bamoun), the frequency information for the amalgamated “*Bantu speakers*” sample was used to replace the missing data. These five populations do not seem to be very different from the point of view of our genetic or linguistic data (paired samples *t* tests between all pairs of these populations, separately for the linguistic and genetic data, are nonsignificant), and, moreover, they do not differ genetically from the “*Bantu speakers*” sample (paired *t* tests are also nonsignificant). These results allow the amalgamation procedure, even if the demographic and linguistic histories of these five populations are very different (1, 2). This procedure could introduce a bias toward those linguistic features uniform across the sampled Bantu languages and against those showing variation. To control for this possibility, two artificial variants, *ASPM-D\** and *MCPH-D\**, were created from *ASPM-D* and *MCPH-D*, by replacing their actual frequency values in the five Bantu populations with their averages. Systematic checks during all stages of the analysis suggest that this missing-data-handling procedure did not unduly distort the results.

The final database comprises 983 alleles, with an unbiased distribution across the chromosomes. For each linguistic feature, the number of alleles correlating with it in the top 5% of the empirical distribution across the chromosomes does not deviate from the expected distribution ( $\chi^2$  tests are nonsignificant), suggesting that there are no chromosomes tending to correlate better with the linguistic features.

**Linguistic Data.** Of the 141 linguistic features in ref. 71, 24 were retained. The criteria for retention were good coverage of the 49 populations and meaningful binary coding. Two new features (*Coda* and *OnsetClust*) were added. The 26 binary linguistic features, covering varied aspects of phonology and morpho-syntax, are as follows: *ConsCat* (are there >25 consonants?), *VowelsCat* (are there >6 vowels?), *UvularC* (are there uvular consonants?), *GlotC* (are there glottalized consonants?), *VelarNasal* (are there velar nasals?), *FrontRdV* (are there front rounded vowels?), *Coda* (are codas allowed?), *OnsetClust* (are onset clusters allowed?), *WALS-SylStr* (is syllable structure at least moderately complex as defined in ref. 71?), *Tone* (does the language have a tonal system?), *RareC* (does the language have any rare consonants?), *Affixation* (does the language use affixes?), *CaseAffixes* (are cases marked with affixes?), *NumClassifiers* (does the language have numeral classifiers?), *TenseAspect* (are there inflections marking tense-aspect?), *MorphImpv* (are there dedicated morphological categories for second person imperatives?), *SVWO* [what is the dominant subject-verb word order (if any)?], *OVWO* [what is the dominant object-verb word order (if any)?], *AdposNP* [what is the dominant order (if any) between adposition and noun phrase?], *GenNoun* [what is the dominant order (if any) between genitive and noun?], *AdjNoun* [what is the dominant order (if any) between adjective and noun?], *NumNoun* [what is the dominant order (if any) between numeral and noun?], *InterrPhr* (are “WH” question words phrase-initial?), *Passive* (is there a passive construction?), *NomLoc* (are locational predication and nominal predication encoded the same way?), and *ZeroCopula* (is omission of copula allowed?).

For each of the 49 populations, the values of these 26 linguistic features were collected. The attribution of values to these features was based, where possible, on published material (71–84), but we also gathered primary data by sending standardized questionnaires to specialists in several of the languages concerned (see acknowledgments). In most instances, this attribution is straightforward, but in some it involves a certain degree of subjective judgment, whereas in some others the data are simply unavailable. Nevertheless, we are confident that most linguists would agree with the vast majority of our decisions.

**Typological Linguistic Distance.** For any set of linguistic features,  $f_i$ , and pair of populations,  $p_1$  and  $p_2$ , the typological linguistic distance is defined as:

$$D_L(f_1, \dots, f_n; p_1, p_2; w_1, \dots, w_n) = \sqrt{\left( \sum w_i (f_{1i} - f_{2i})^2 \right)}$$

The equal weighting scheme considers all features equally important:  $w_1 = \dots = w_n = 1/n$ . Let  $H_i$  be the informational entropy (85) of linguistic feature  $f_i$ ; then the direct proportion weighting scheme considers more important those features that carry more information,  $w_i = H_i/\sum H_i$ , whereas the inverse proportion weighting scheme considers more important those features whose distribution is more skewed,  $w_i = 1/[H_i \sum (1/H_i)]$ . These three weighting schemes intercorrelate extremely well (Mantel's  $r = 0.996, 0.978$ , and  $0.959$ , respectively,  $P < 0.001$ ), so that only the equal weighting scheme was used.

We thank B. Connell, C. Kutsch Lojenga, H. Eaton, J. A. Edmondson, J. Hurford, K. Bostoen, L. Ziwo, M. Blackings, N. Fabb, O. Stegen, R. Asher, R. Ridouane, M. Endl, and J. Roberts for help with language data; A. Dima for help with statistics; and J. Hurford, S. Kirby, R. McMahon, S. Della Sala, T. Bates, and P. Wong for discussions and comments. We also thank three anonymous reviewers for their suggestions. D.D. was funded by an Overseas Research Students Award and a Studentship from the College of Humanities and Social Science, University of Edinburgh. D.R.L. acknowledges the support of an Individual Research Fellowship from the Leverhulme Trust.

<sup>8</sup>Giving the three-letter language codes (4). These linguistic attributions are not unique in some cases.

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